Loss and Optimization

Loss Functions

How "bad" is our model?

$$L(\theta) = \sum_{i} \mathcal{L}(\theta; x_{i}, y_{i})$$

Example: least-squares regression

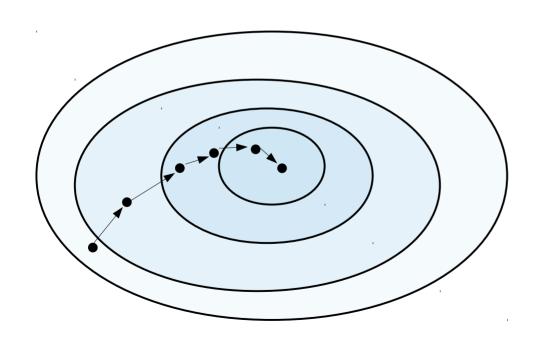
$$L(\theta) = \sum_{i} (y_{i} - f(\theta; x_{i}))^{2}$$

Classification: Negative log-likelihood

$$L(\theta) = \sum_{i} -\log[p(y_{i}|\theta,x_{i})]$$

Optimization – Gradient Descent

How do we find $\operatorname{argmin}_{\theta}L(\theta)$



Initialize random θ

N iterations:

Compute $abla_{\scriptscriptstyle{ heta}} L(heta)$

$$\theta \leftarrow \theta - \epsilon \nabla_{\theta} L(\theta)$$

Gradients

• How do we compute $\nabla_{\theta} L(\theta)$?

$$\nabla_{\boldsymbol{\theta}} L(\boldsymbol{\theta}) = \nabla_{\boldsymbol{\theta}} \Big[\sum_{i} \mathcal{L}(\boldsymbol{\theta} \mid \boldsymbol{x}_{i}, \boldsymbol{y}_{i}) \Big] = \sum_{i} \nabla_{\boldsymbol{\theta}} \mathcal{L}(\boldsymbol{\theta} \mid \boldsymbol{x}_{i}, \boldsymbol{y}_{i})$$

Squared error:
$$\mathcal{E}(\theta \mid x_i, y_i) = \left[(\mathbf{W} \mathbf{x}_i + b) - y_i \right]^2$$
$$\nabla_{\mathbf{W}} \left[(\mathbf{W} \mathbf{x}_i + b) - y_i \right]^2 = 2 \left[(\mathbf{W} \mathbf{x}_i + b) - y_i \right] \mathbf{x}_i$$

$$\nabla_{\mathbf{b}} [(\mathbf{W} \mathbf{x}_i + b) - y_i]^2 = 2[(\mathbf{W} \mathbf{x}_i + b) - y_i]$$

Negative log-likelihood: $\mathcal{E}(\theta \mid x_i, y_i) = -\log(\operatorname{softmax}(W x_i + b))_{y_i}$

$$\nabla_{\mathbf{W}_{i}} \left(-\log\left(\operatorname{softmax}\left(\mathbf{W}\,\mathbf{x}_{i} + \mathbf{b}\right)\right)_{y_{i}}\right) = \left(\operatorname{softmax}\left(\mathbf{W}\,\mathbf{x}_{i} + \mathbf{b}\right)_{j} - \left[y_{i} = j\right]\right)\mathbf{x}_{i}$$

$$\nabla_{\mathbf{b}_{i}} \left(-\log\left(\operatorname{softmax}\left(\mathbf{W}\,\mathbf{x}_{i}+\mathbf{b}\right)\right)_{y_{i}}\right) = \operatorname{softmax}\left(\mathbf{W}\,\mathbf{x}_{i}+\mathbf{b}\right)_{j} - \left[y_{i}=j\right]$$

Gradients – Computation Graphs

$$(\mathbf{y} - (\mathbf{W} \mathbf{x} + \mathbf{b}))^2$$

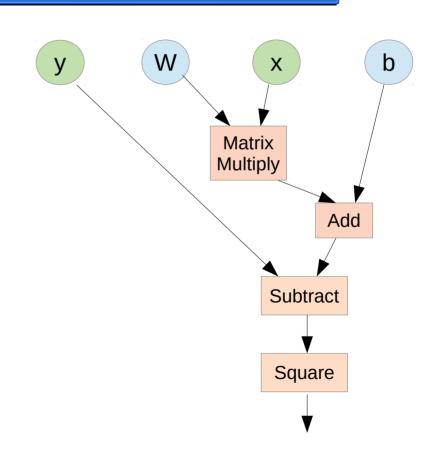
sq(sub(y, add(matmul(W, x), b)))

 $\frac{\partial}{\partial W}$ sq(sub(y, add(matmul(W, x), b)))

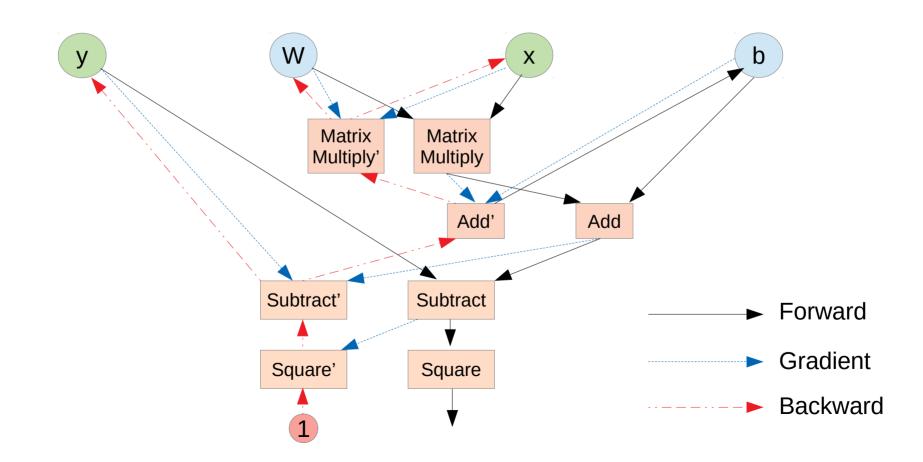
= sq'(sub(y, add(matmul(W, x), b))) * $\frac{\partial}{\partial W}$ sub(y, add(matmul(W, x), b))

=...

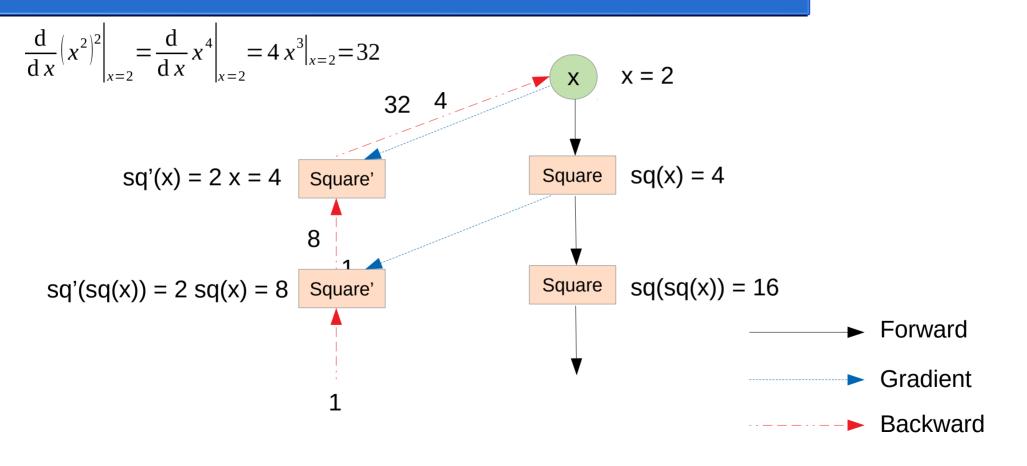
= sq'(sub(y, add(matmul(W, x), b))) *
sub'(y, add(matmul(W, x), b)) *
add'(matmul(W, x), b) *
matmul'(W, x)



Gradients – Computation Graphs



Gradients – Computation Graphs



Learning Rate

$$\theta \leftarrow \theta - \epsilon \nabla_{\theta} L(\theta)$$

